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Model-Based Algorithms for Detecting Cable Damage from Time-Domain Reflectometry Measurements

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May 7, 2007

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Albuquerque, NM, United States
May 7, 2007 through May 10, 2007

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Model-Based Algorithms for Detecting Cable Damage from Time-Domain Reflectometry Measurements

GRACE A. CLARK

NSED, SYSTEMS AND DECISION SCIENCES SECTION

MAY 2, 2007

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Agenda

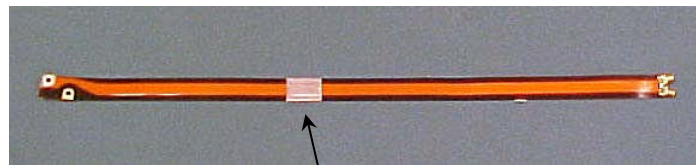


- Introduction and Problem Definition - *Work in Progress*
- Technical Approach - *Model-Based Damage Detection*
- Model-Based Damage Detection Results
- Discussion

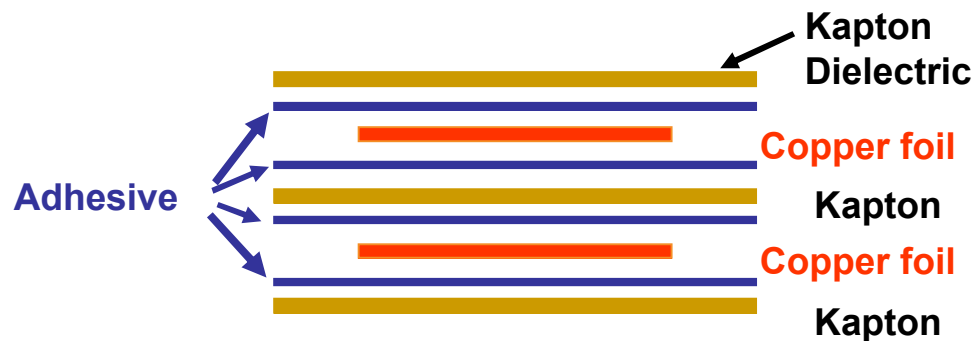
We Are Testing Two-Conductor Flat Cables With Kapton Insulation



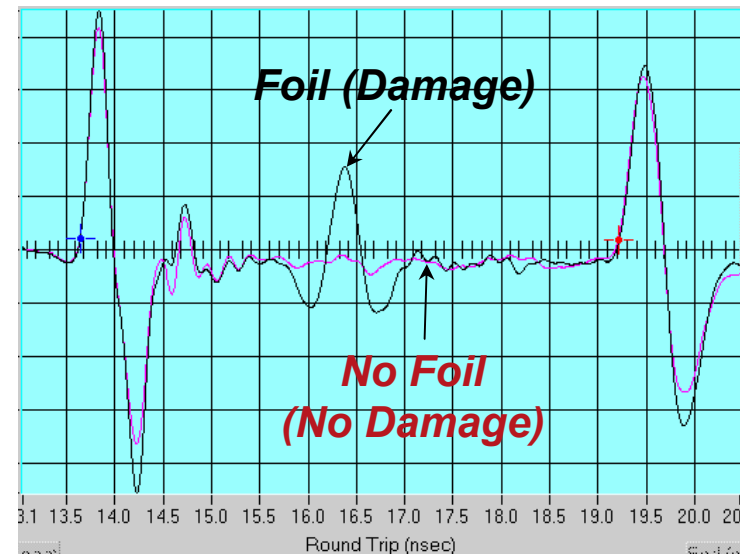
**Two-Conductor Flat Cable
With Kapton Insulation**



**Foil Simulating a Capacitive
Discontinuity (Damage)**

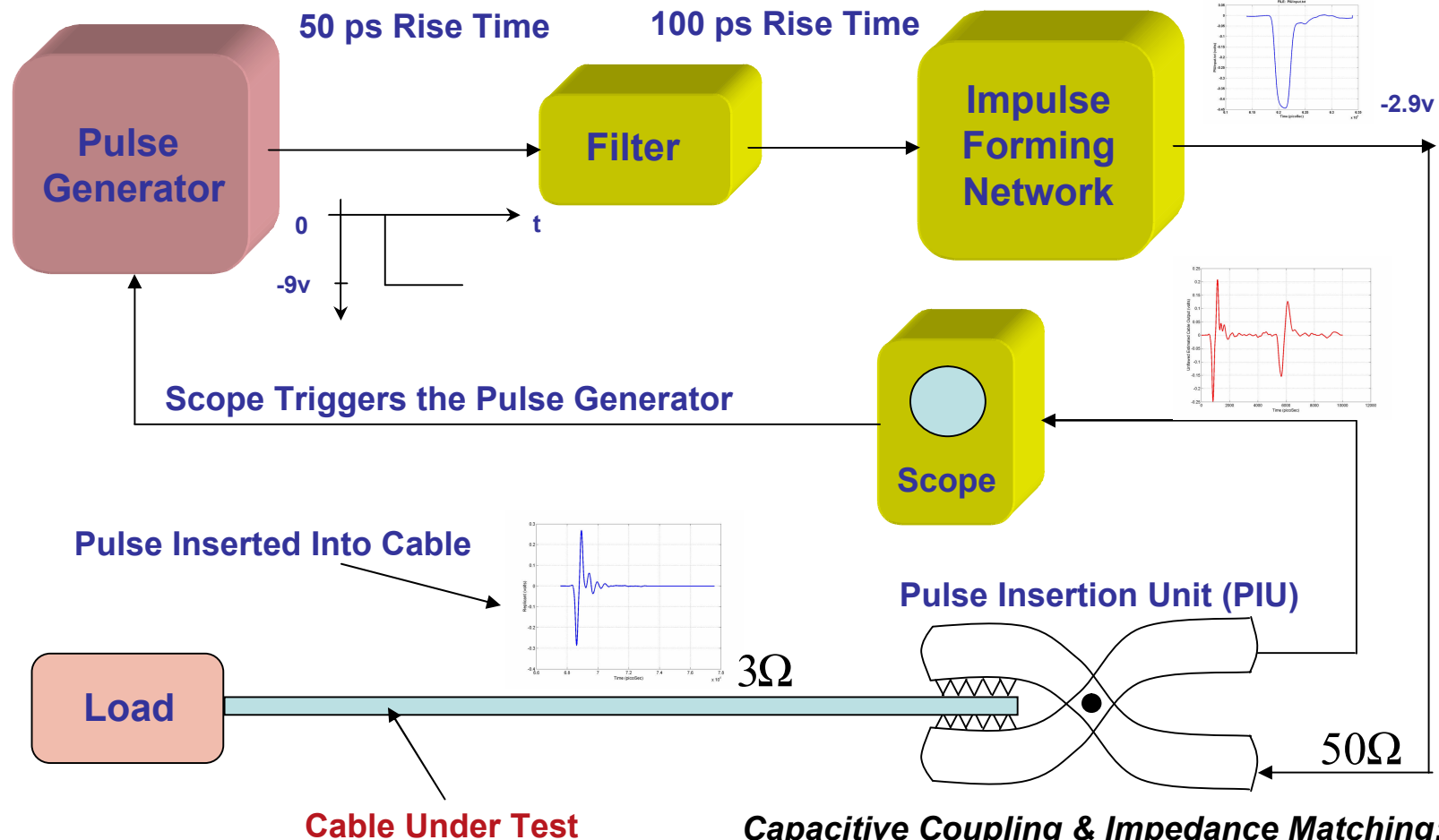


Red TDR Signal => Good Cable
Black TDR Signal => Damaged Cable



Benchtop Experiments (w/No Device “Mockup):” Connections Create Some Variability

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Capacitive Coupling & Impedance Matching:

- PIU = Half of “The Capacitor”
- Cable = Half of “The Capacitor”

Proposed Decision-Making Protocol (Using TDR Measurements):



Use a Three-Step Hierarchical Decision Scheme:

1. Detection:

- *Decide whether or not an abnormality in the cable TDR response exists (yes or no)*
- *Assume that an abnormal TDR response implies a flaw in the cable*

2. Flaw or Failure Mode Classification:

- *Classify the type of failure mode or flaw detected, from among a fixed set of possible modes*

3. Final Decision:

- *Using all of the information from the measurements and the previous two steps (fusion), decide whether the cable is “reliable or not reliable”*

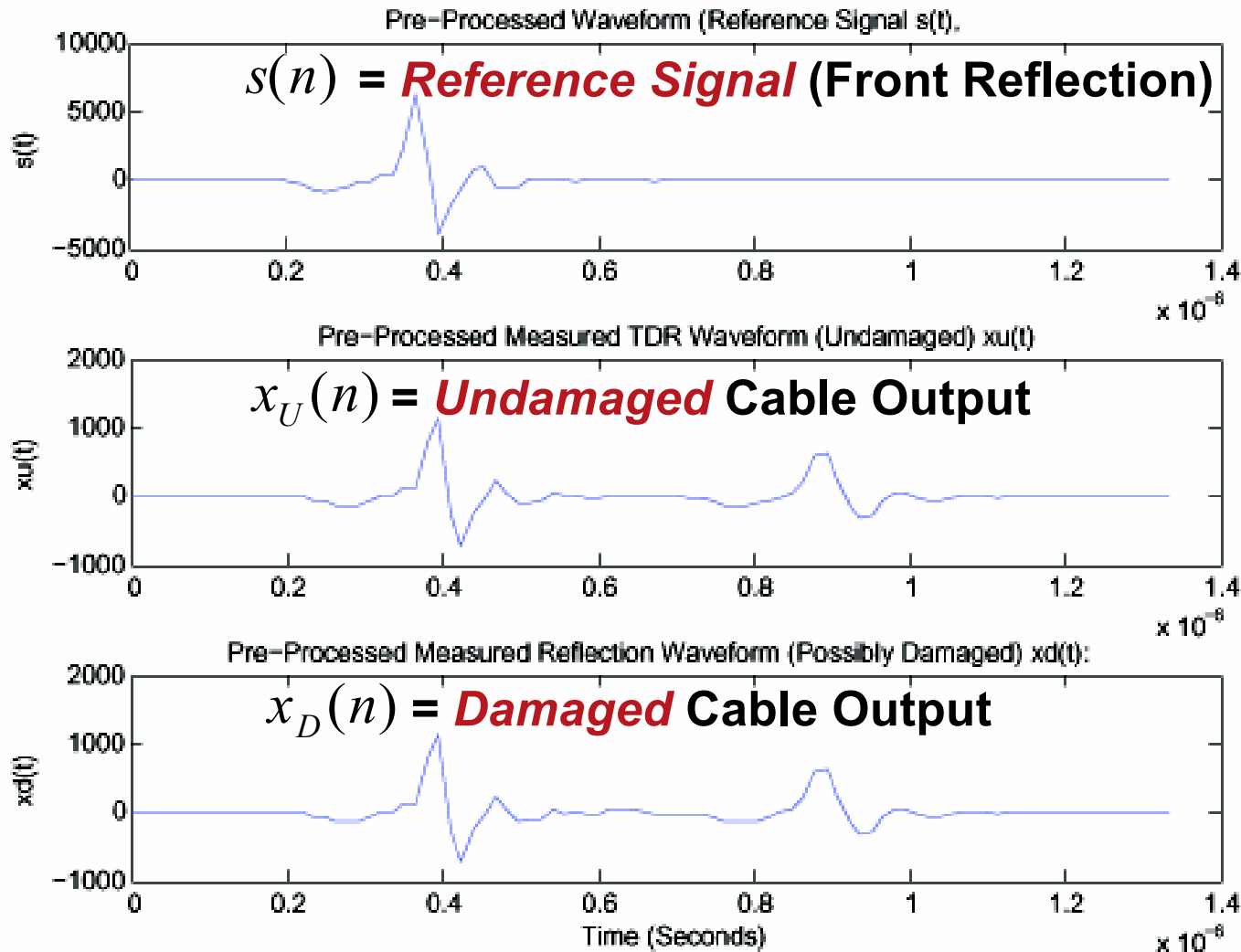
Model-Based Damage Detection: Detect a Model Mismatch if a Damage is Present



- Exploit the fact that the TDR measurements are reasonably repeatable.
- Build a forward model of the dynamic system (cable) for the case in which *NO DAMAGE* exists
- Whiteness Testing on the *Innovations (Errors)*:
Estimate the output of the actual system using measurements from a dynamic test.
 - If *no damage* exists, the model will match the measurements, so the “innovations” (errors) will be *statistically white*.
 - If a *damage* exists, the model will not match the measurements, so the “innovations” (errors) will *not be statistically white*.
- Weighted Sum Square Residuals (WSSR) Test:
The WSSR provides a single metric for the model mismatch

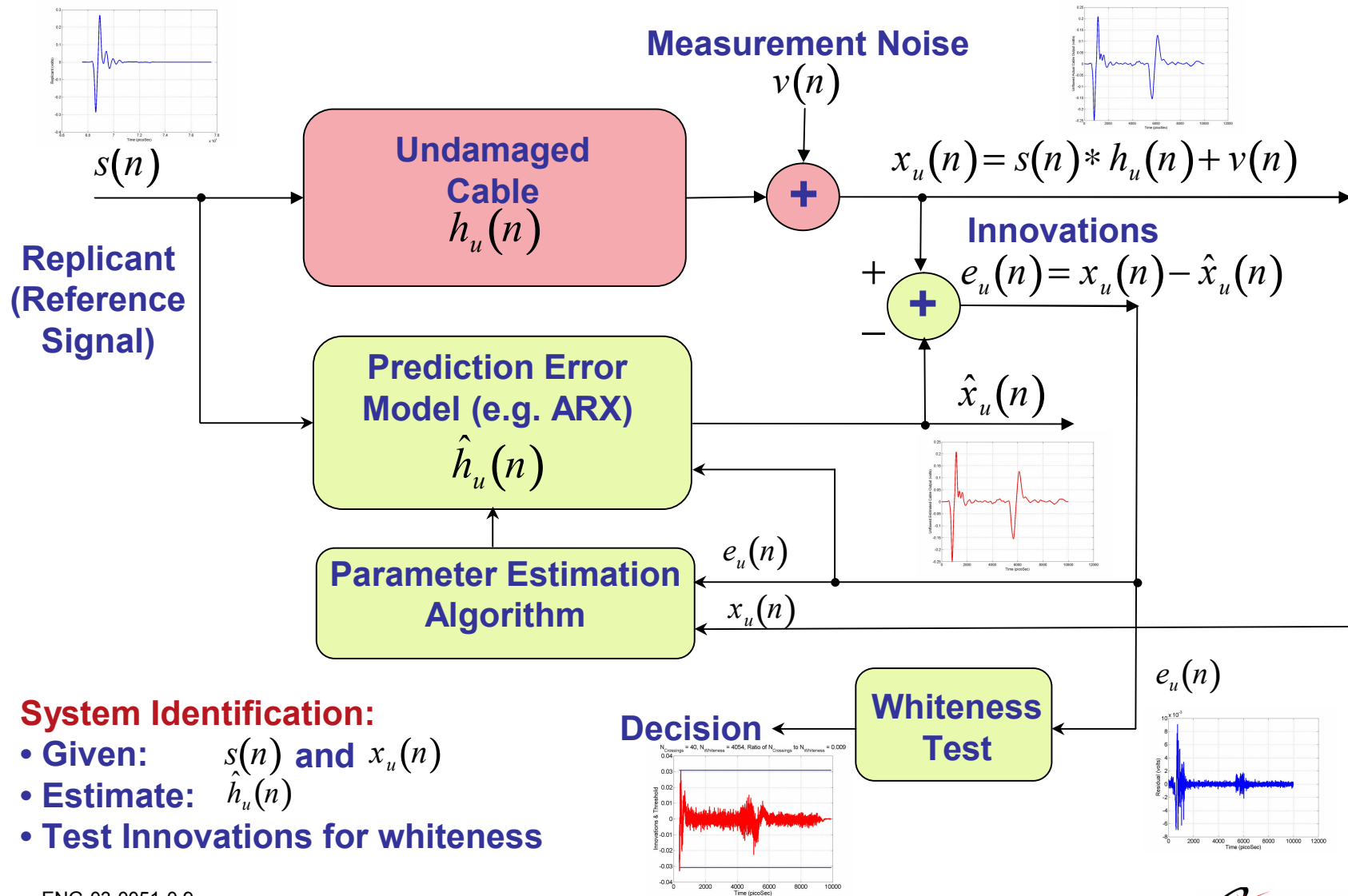
Experiment Using Real Cable TDR Signals: Pre-Processed Measurements

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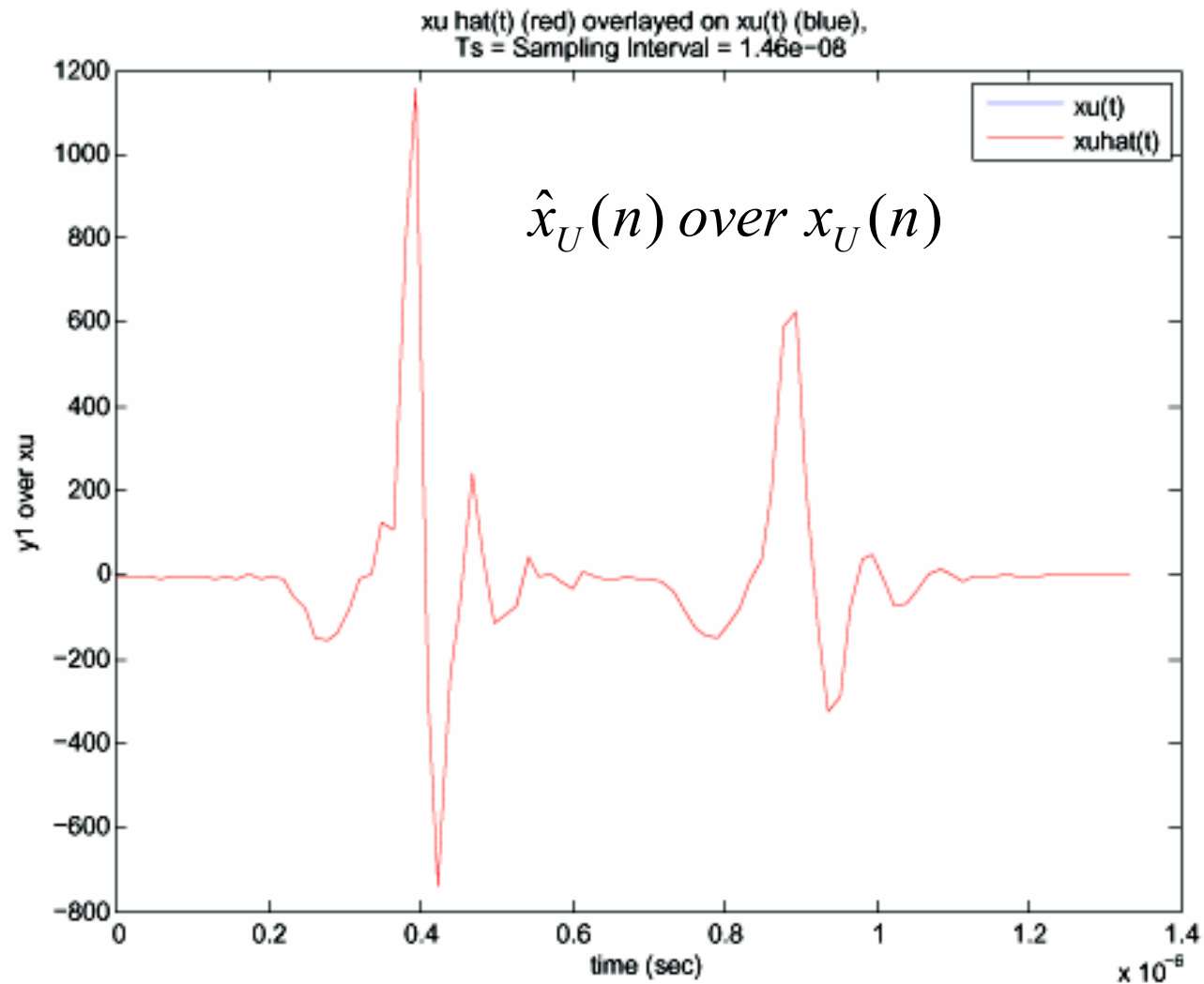
Step #1: System Identification to Estimate the Dynamic Model of the *Undamaged Cable*

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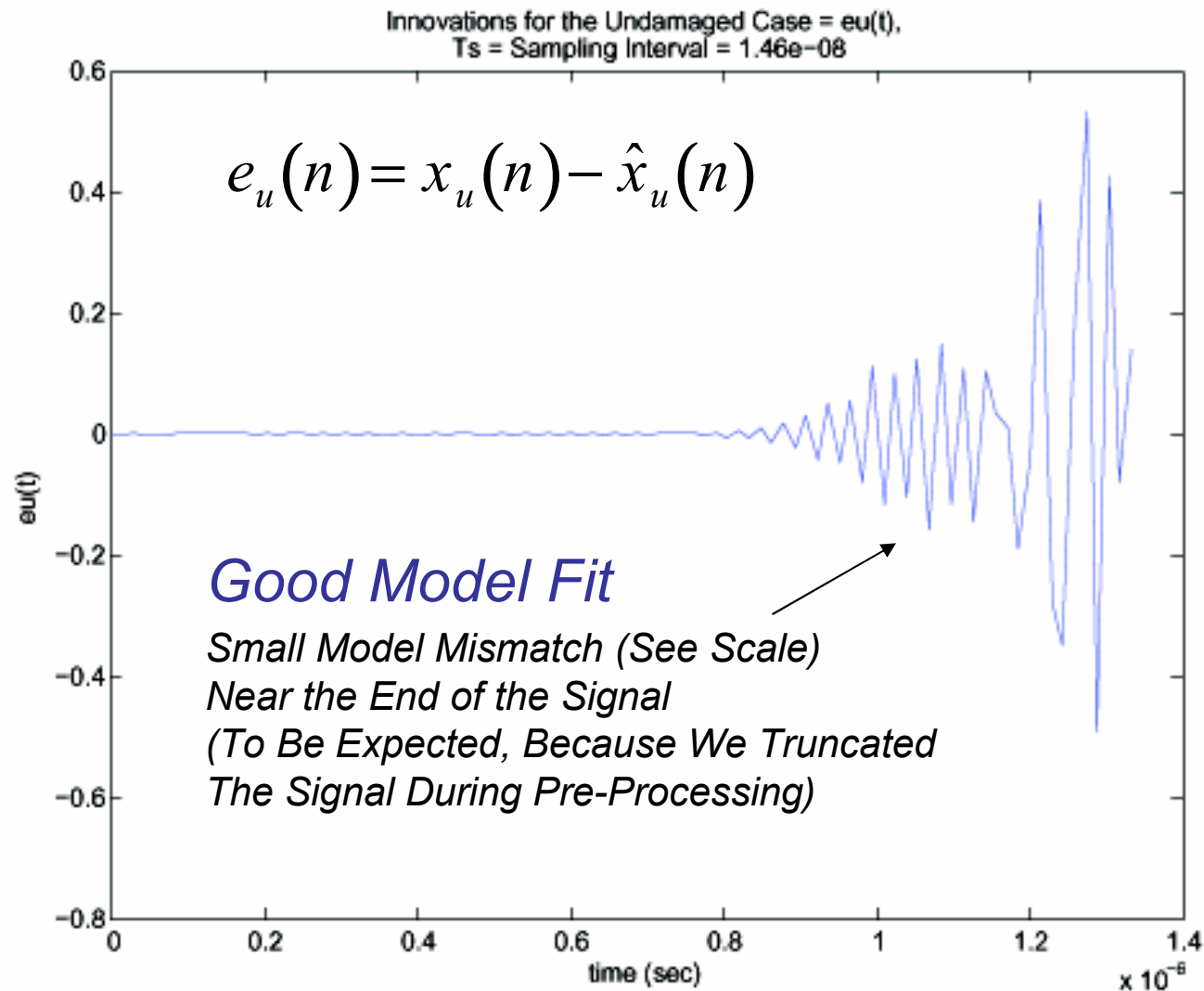
Undamaged Case:

$\hat{x}_U(n)$ **Plotted Over** $x_U(n)$ (Good Model Fit)



Undamaged Case:

$e_u(n)$ = Residual (or “Innovations”)



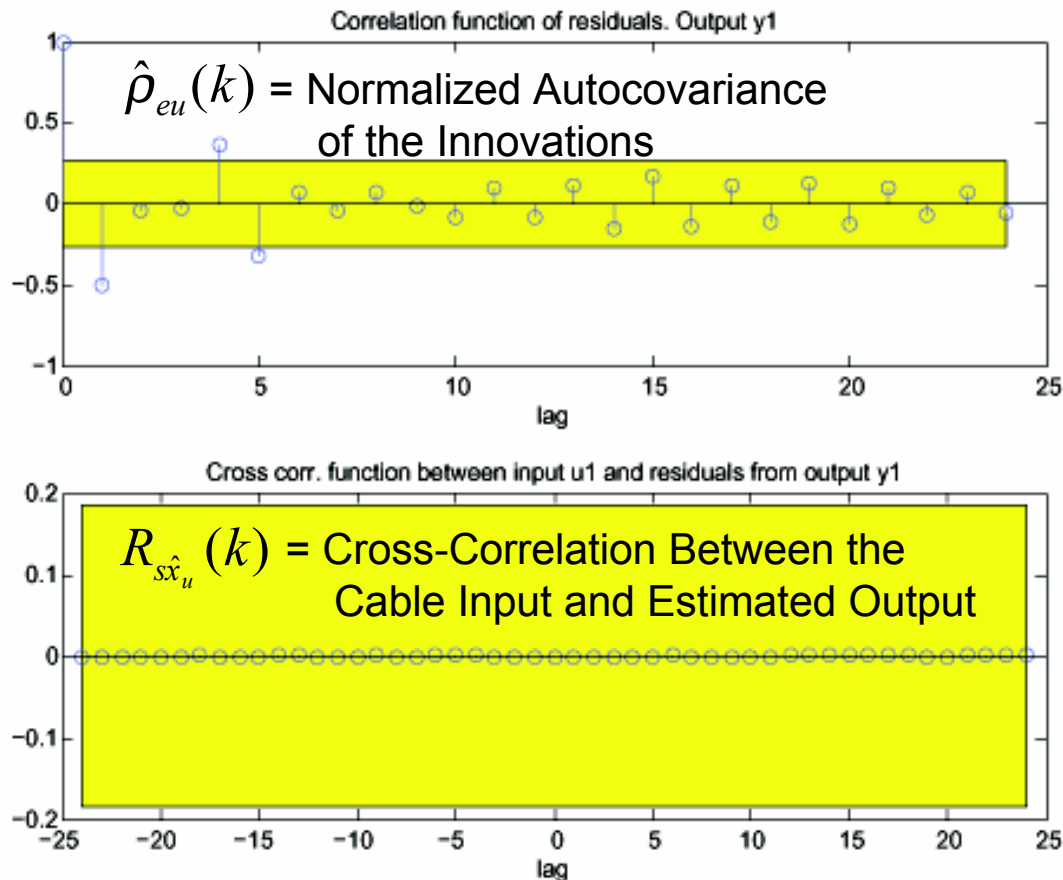
Undamaged Case:

Whiteness Test on the Innovations $e_u(n)$

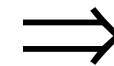
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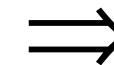
$$e_u(n) = x_u(n) - \hat{x}_u(n) = \text{Innovations}$$



- The innovations pass the Whiteness Test
- The Cross-Correlation is Very Small



Declare that the Innovations are “White”



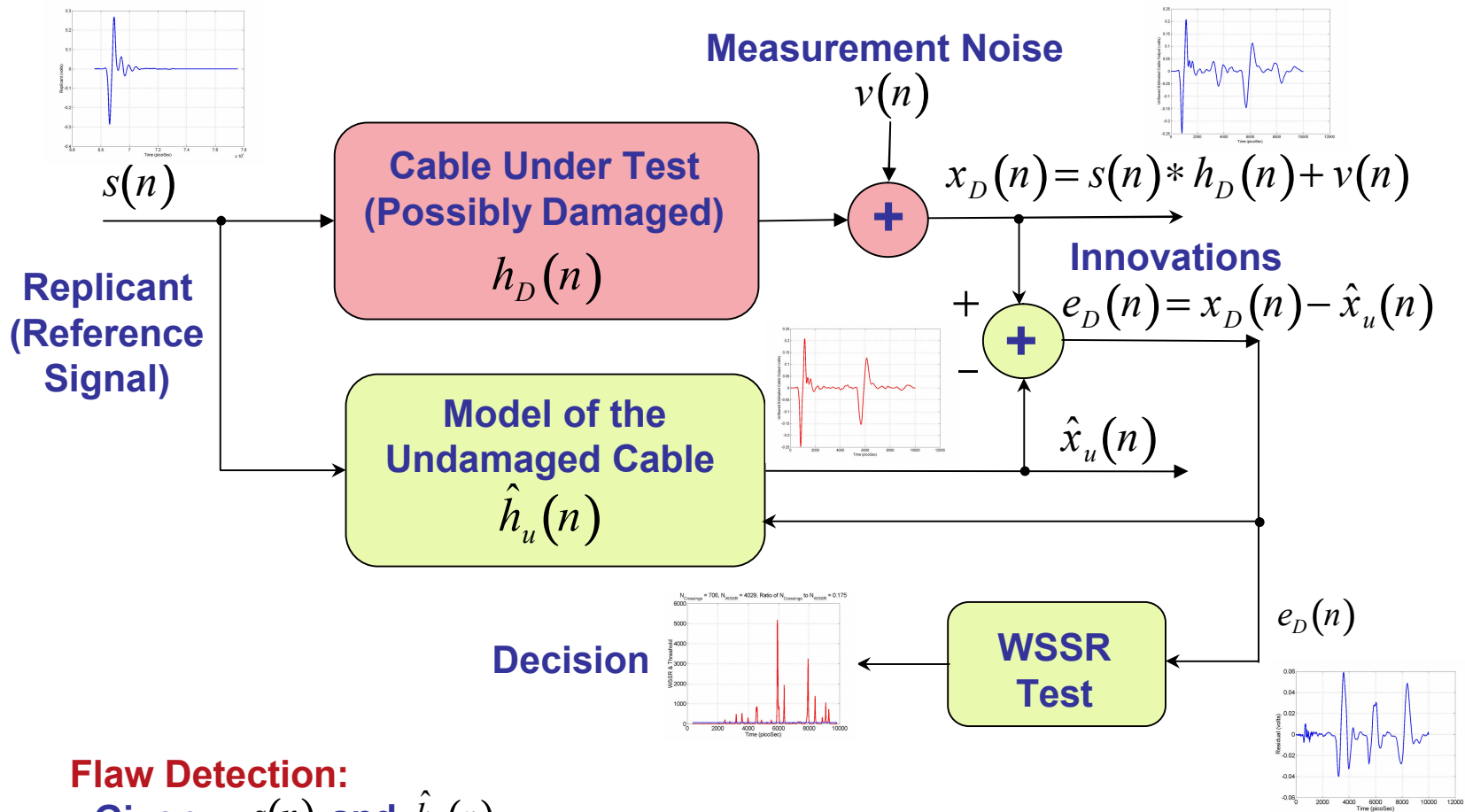
There is no model mismatch



The model is valid

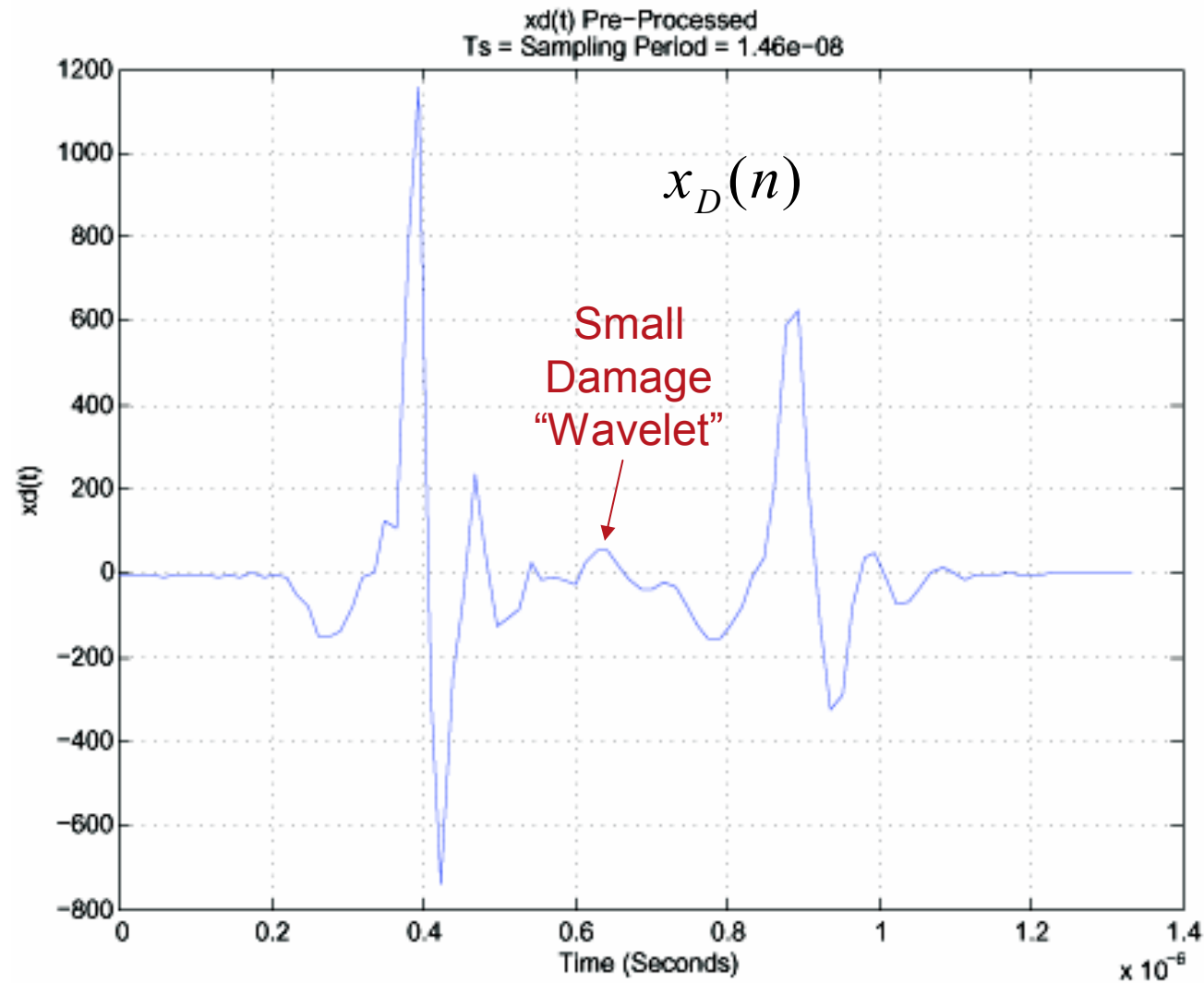
Step #2: Compare the Responses of the Undamaged and Damaged Cables ==> Damaged Detection

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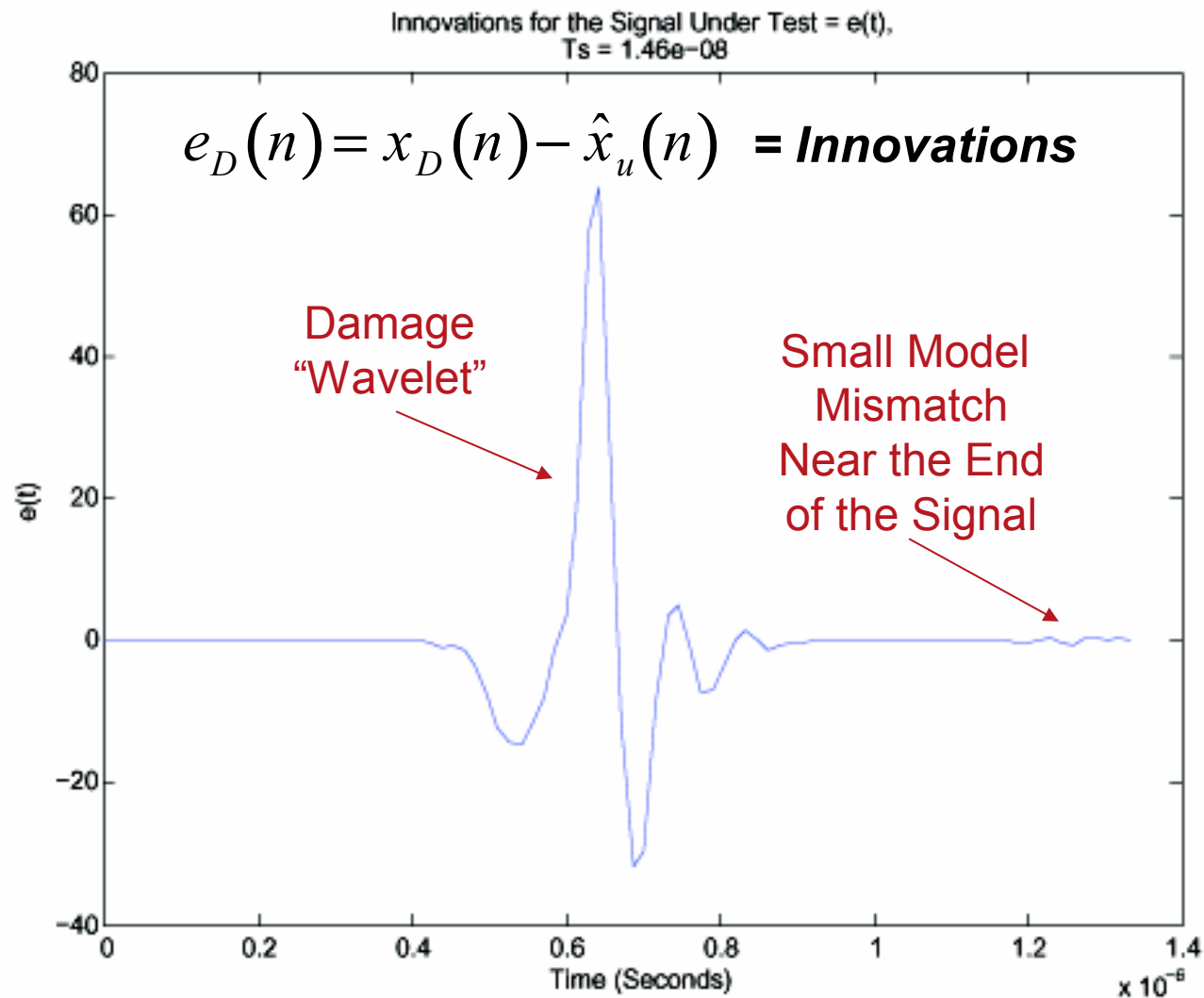
Damaged Case:

$x_D(n)$ = **Damaged** Cable Output



Damaged Case:

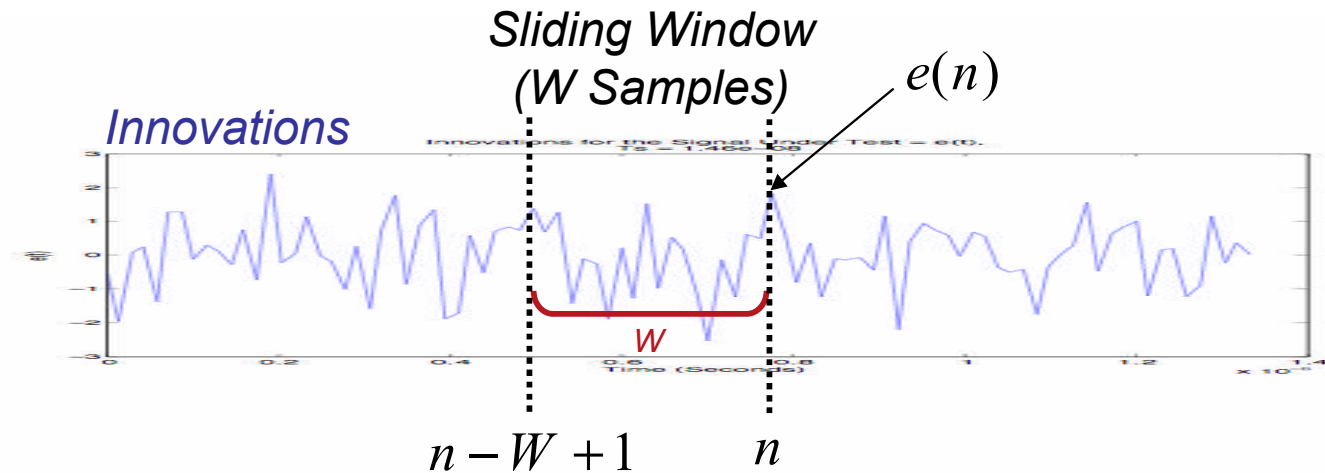
$e_D(n)$ = Residual (or “Innovations”)



WSSR is Calculated Using a Sliding Window Over the Innovations Sequence $e(n)$



WSSR = “Weighted Sum Squared Residuals”



$$\gamma(n) = \sum_{j=n-W+1}^n \frac{e^2(j)}{V(j)}, \quad \text{for } n \geq W$$

WSSR is a useful test statistic for detecting an abrupt change, or “jump” in the innovations

The Scalar WSSR Test (Continued)



Summary of the WSSR Test for Significance $\alpha = .05$:

$$\gamma(n) = \sum_{j=n-W+1}^n \frac{e^2(j)}{V(j)}, \quad \text{for } n \geq W$$

$$V(n) = \frac{1}{W} \sum_{j=n-W+1}^n [e^2(j) - \bar{e}(j)]^2, \quad \text{for } n \geq W$$

$$\bar{e}(n) = \frac{1}{W} \sum_{j=n-W+1}^n e(j), \quad \text{for } n \geq W$$

$$\tau = W + 1.96\sqrt{2W}$$

If $\gamma(n) \begin{matrix} \geq H_1 \\ < H_0 \end{matrix} \tau$, $(\tau = \text{Decision Threshold})$

In practice, we implement the WSSR test as follows:

- Let F_E = Fraction of samples of $\gamma(n)$ that exceed the threshold
- If $F_E \leq \alpha$, Declare H_0 is true (innovations are white, no jump)
- If $F_E > \alpha$, Declare H_1 is true (innovations are not white, jump)

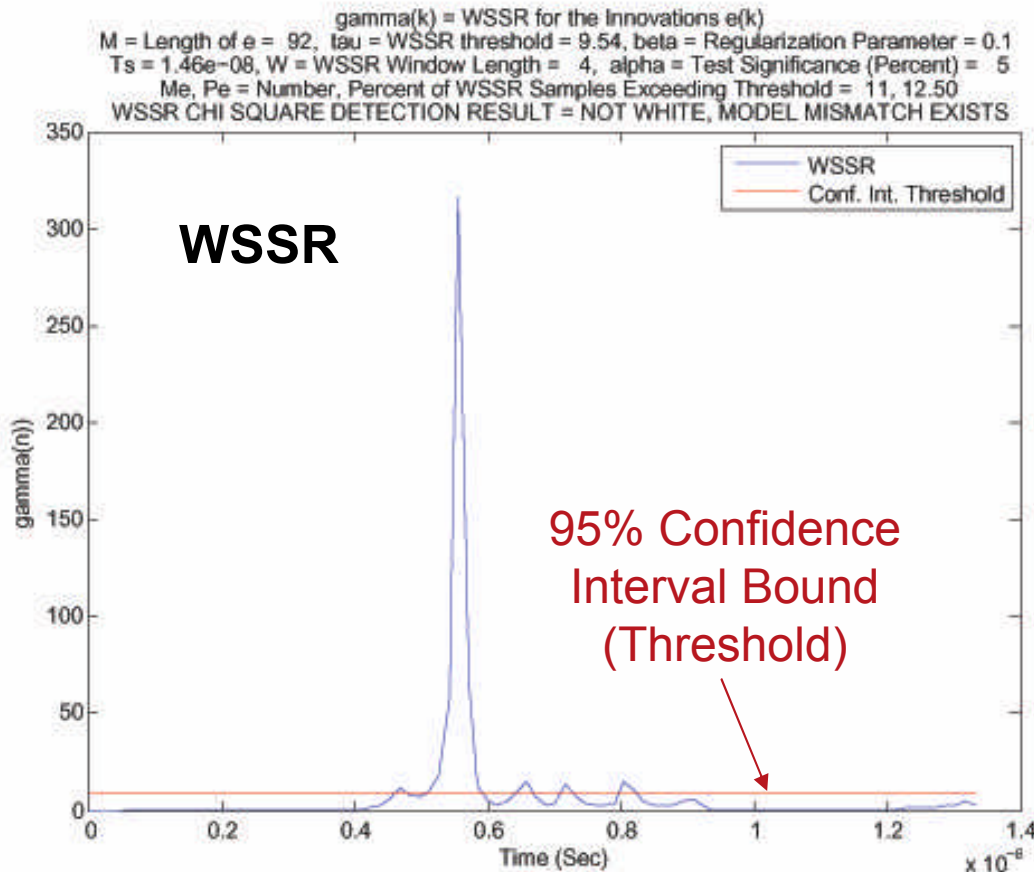
Damaged Case: WSSR Test For the Damaged Case

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$$\gamma(n) = \sum_{j=n-W+1}^n \frac{e^2(j)}{V(j)}, \quad \text{for } n \geq W$$

**WSSR = Weighted Sum
Squared Residuals**



The Innovations Fail the
WSSR Test

> 5% of Samples Exceed
Threshold

⇒ There exists a
model mismatch

⇒ The undamaged model is
NOT Valid for this
cable

⇒ **An anomaly exists
in the cable**

Discussion: The Model-Based Approach Offers Advantageous Properties

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- We can estimate the *LOCATION* of any detected anomaly.
- The algorithm is *robust* with respect to variations in the measured signals for various experimental scenarios:

==> If the TDR signals vary for various scenarios, we can model each case and test the cables effectively.
- This algorithm is very effective, even if we are given *only a single exemplar* of an undamaged cable signal.

Discussion: Future Work:

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- Thorough repeatability studies:
 - Measurement-to-measurement for one cable
 - Cable-to-cable
- Given ensembles of measurements,
we can build more extensive performance measures:
 - Receiver Operating Characteristic (ROC) curves
Probability of Detection
vs. Probability of False Alarm
 - Statistical Confidence Interval about the estimated
probability of correct classification
- Experiments in a cable environment (not just bench-top)
- Cable “insult” studies using estimated damage types

Contingency VG's



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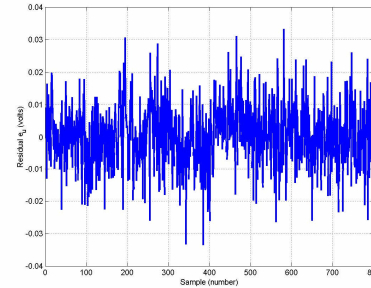
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Let Us Define a “White Noise” Sequence $x(t)$



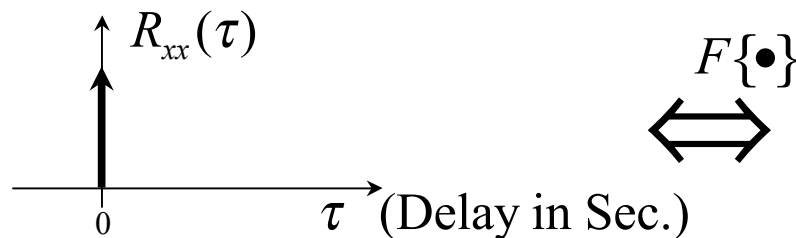
Given a stochastic process $x(t)$



$x(t)$ is “white” when:

Autocorrelation
(Time Domain)

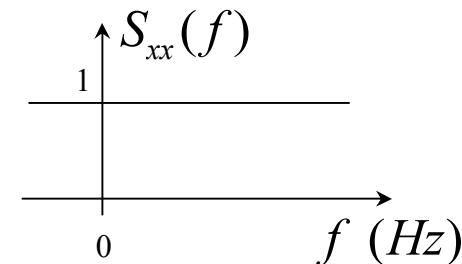
$$\begin{aligned} R_{xx}(\tau) &= E\{x(t)x(t+\tau)\} \\ &= \delta(\tau) \\ &= \begin{cases} 1, & \tau = 0 \\ 0, & \tau \neq 0 \end{cases} \end{aligned}$$



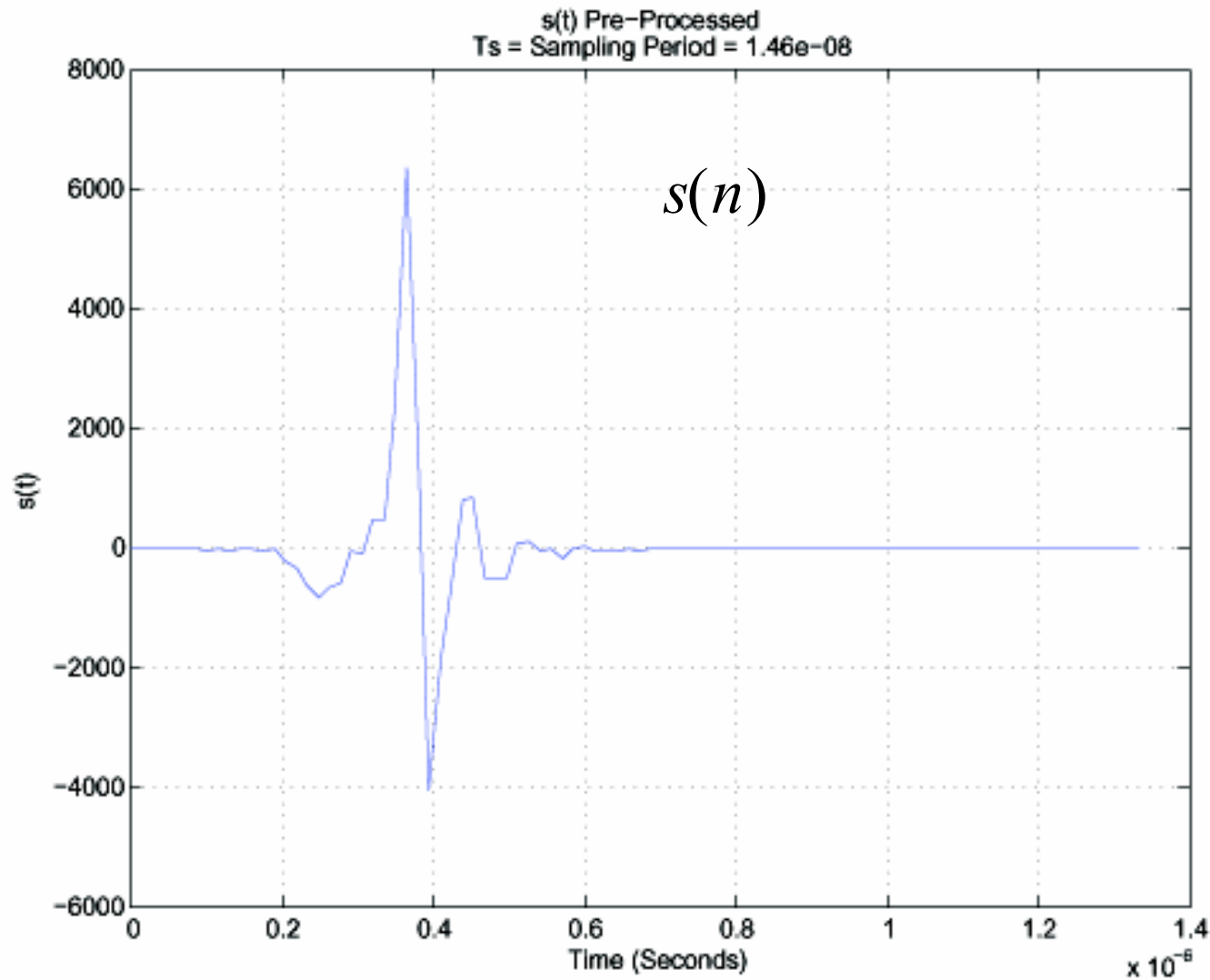
Power Spectral Density
(Frequency Domain)

$$\begin{aligned} S_{xx}(f) &= F\{R_{xx}(\tau)\} \\ &= 1 \end{aligned}$$

$F\{\bullet\}$ = Fourier Transform

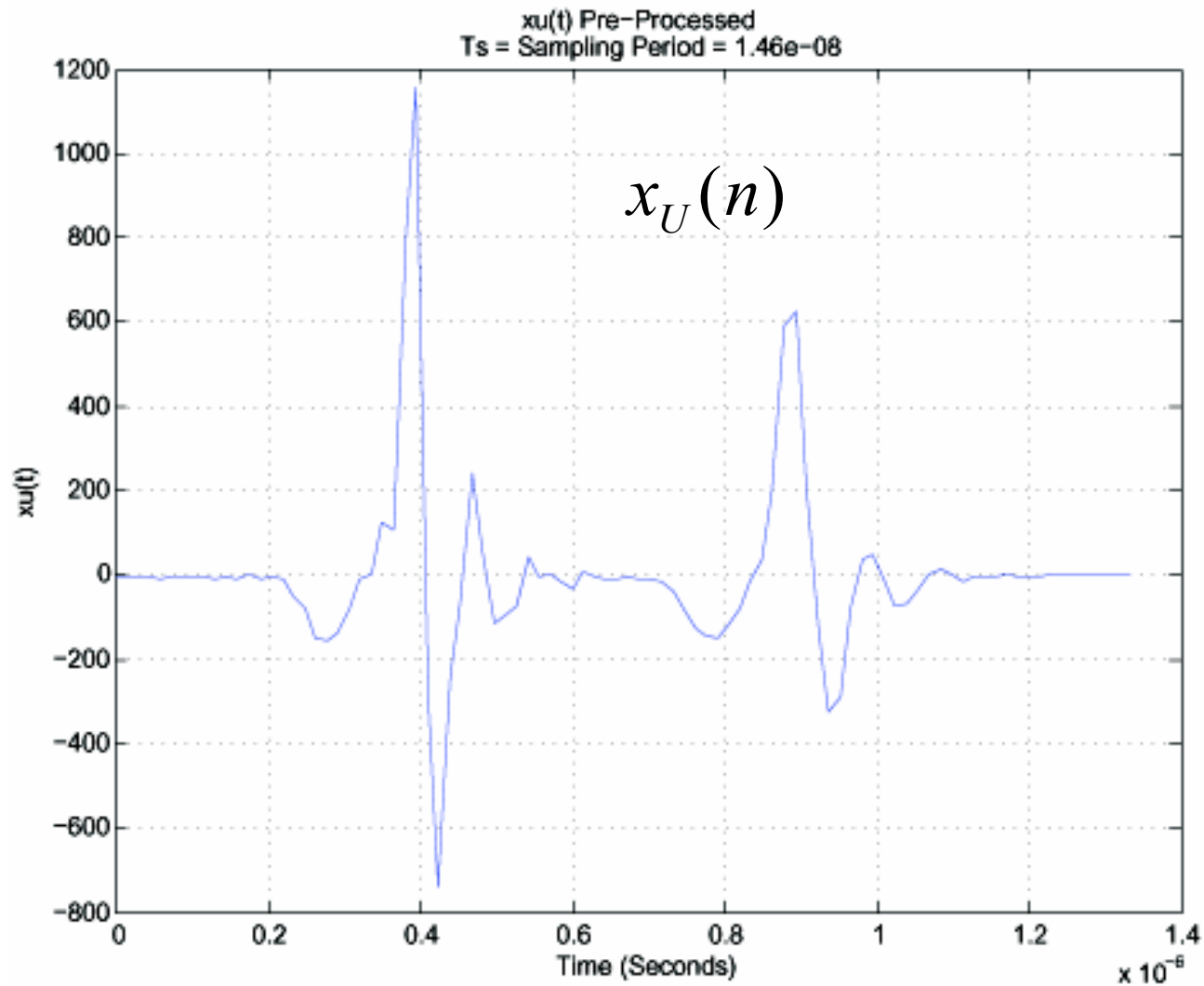


$s(n)$ = *Reference Signal* (Front Reflection)



Undamaged Case:

$x_U(n)$ = **Undamaged** Cable Output

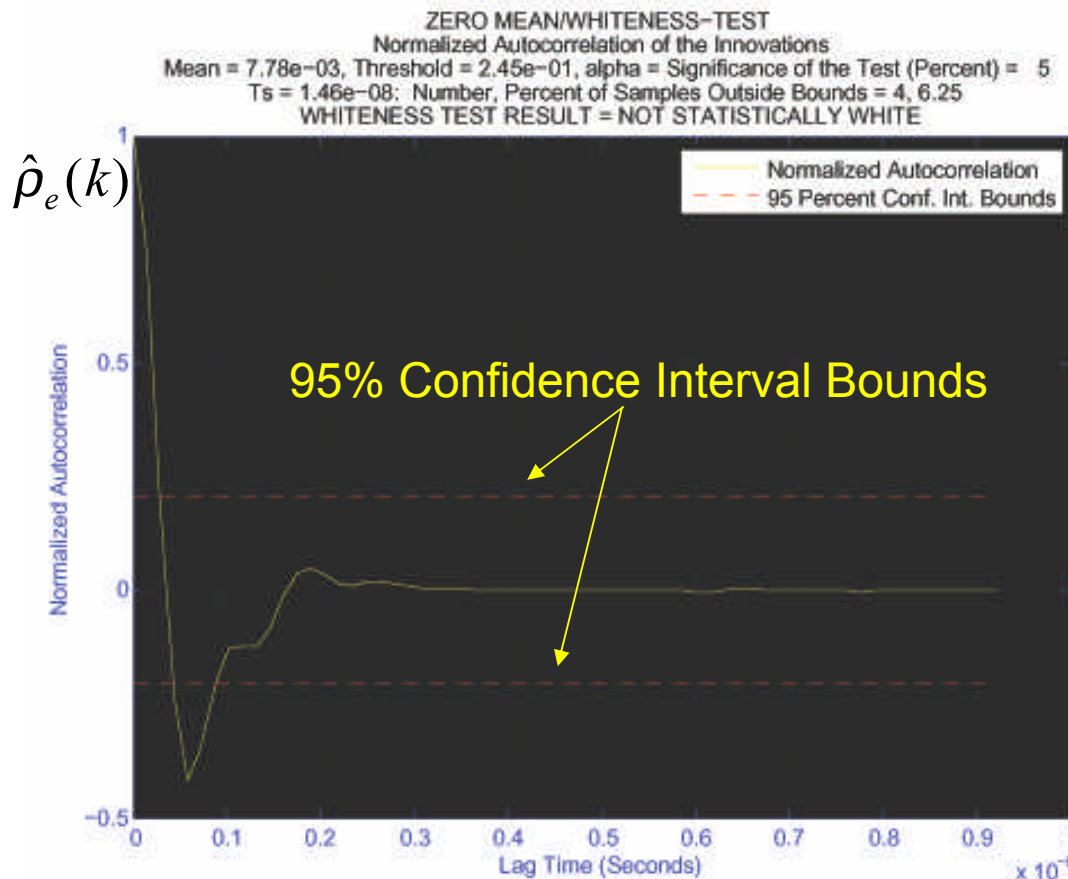


Damaged Case: Whiteness Test For the Damaged Case

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$$e_D(n) = x_D(n) - \hat{x}_u(n) = \text{Innovations}$$



The normalized auto-
Covariance $\hat{\rho}_e(k)$
Does Not Pass
the 95% Confidence
Interval Test

- ⇒ Declare that the Innovations are “Not White”
- ⇒ There exists a model mismatch
- ⇒ The undamaged model is **NOT** Valid for this cable
- ⇒ **An anomaly exists in the cable**

Appendix: System Identification Using an ARMAX Model



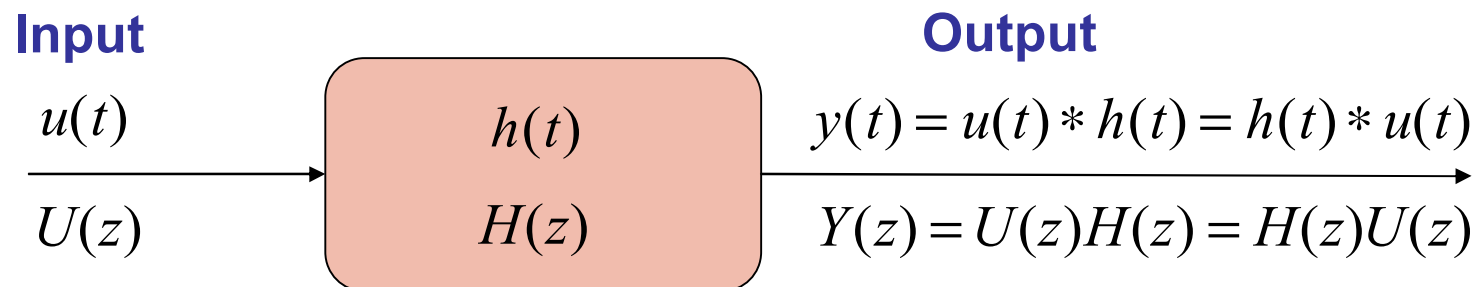
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(Viewgraph in Progress!)

We Can Write the *Impulse Response* $h(t)$ and *Transfer Function* $H(z)$ of a Linear System



- Assume that the system is linear and time-invariant
- Use the discrete-time system representation



- The transfer function can be represented by a rational polynomial in the Z-Transform variable, z

$$H(z) = \frac{B(z)}{A(z)}$$

Where:

$$A(z) = 1 + a_1 z^{-1} + a_2 z^{-2} \mathbf{L} + a_{N_a} z^{-N_a}$$
$$B(z) = b_0 + b_1 z^{-1} + b_2 z^{-2} \mathbf{L} + b_{N_b} z^{-N_b}$$

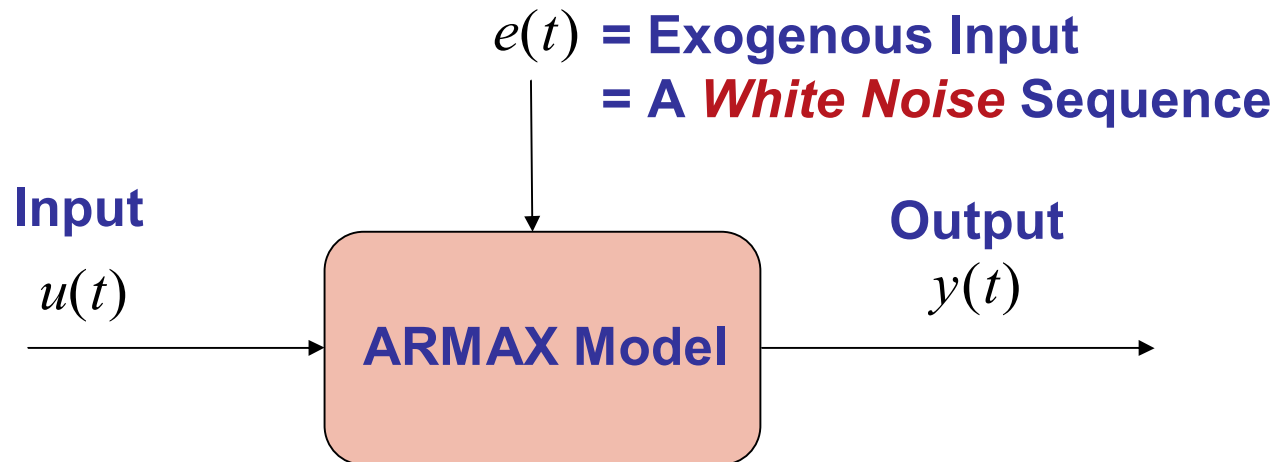
The General Form of the System Model is Called “ARMAX”



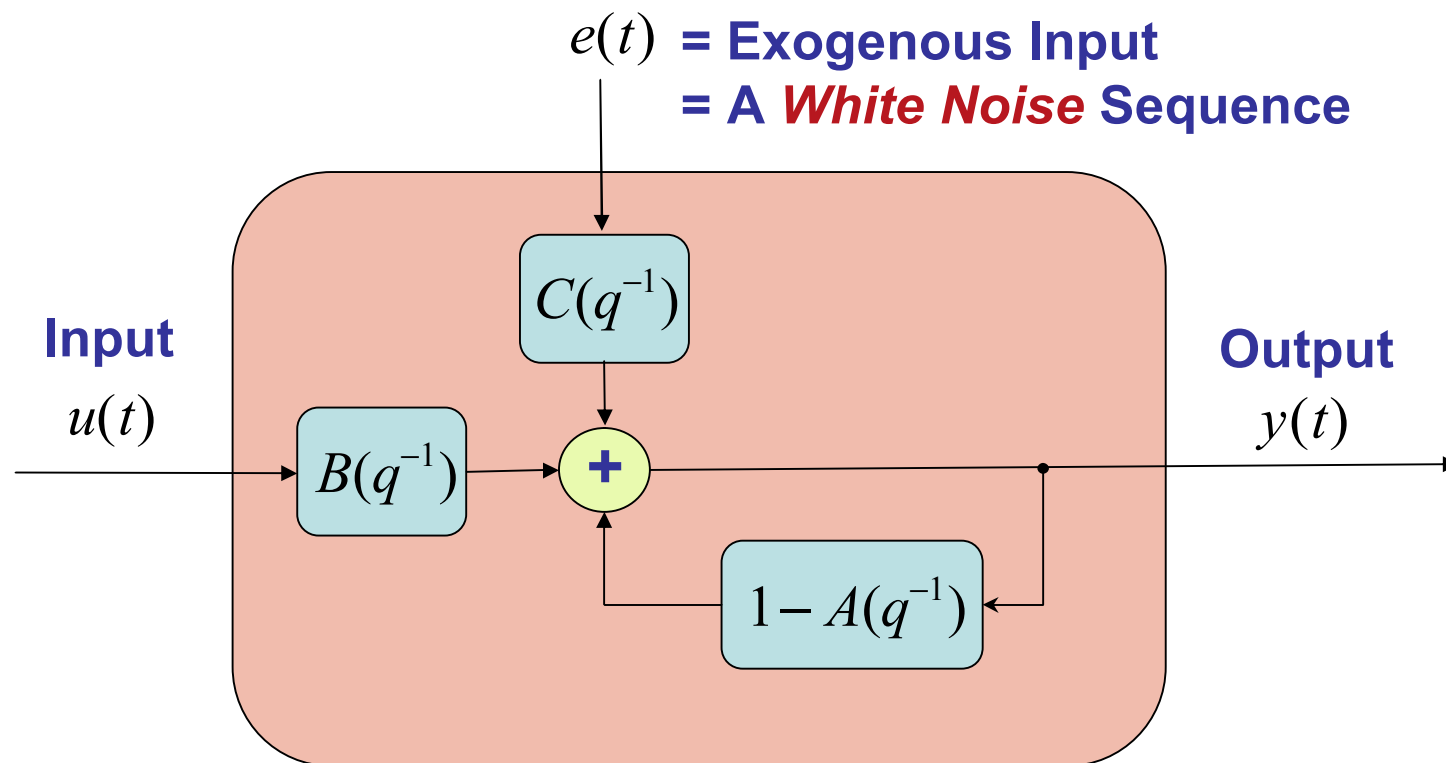
- ARMAX means “Autoregressive Moving Average with Exogenous Input” (“Exogenous” ==> External)
- Let q^{-1} denote the delay operator, so $q^{-k}y(t) = y(t-k)$
- The following model is **ARMAX(Na,Nb,Nc)**:

$$A(q^{-1})y(t) = B(q^{-1})u(t) + C(q^{-1})e(t)$$

Where: $C(z) = 1 + c_1z^{-1} + c_2z^{-2} + \dots + c_{N_c}z^{-N_c}$



We Can Draw A Signal Flow Diagram of the **ARMAX** Model:



WSSR for the Scalar Case

(One Measurement Only, $p = 1$)



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WSSR (Weighted Sum Squared Residuals) Test

For a Scalar Measurement ($p = 1$)



WSSR is used to deal with two main issues:

1) Multiple measurements ($p > 1$):

WSSR allows us to aggregate multiple whiteness tests into a single aggregated scalar test over all of the measurements.

2) Nonstationary Prediction Errors (Innovations):

When the signals are nonstationary, WSSR is a more reliable statistic to use for testing the whiteness of the prediction error sequence (innovations). We require that the WSSR lies beneath a calculated threshold to deem the innovations zero-mean and white.

***WSSR is calculated over a sliding window of W samples.
It is a useful test statistic for detecting an abrupt change in
the innovations signal***

Under the zero mean assumption, the WSSR statistic is equivalent to testing that the prediction error sequence is white.

Scalar WSSR (Weighted Sum Squared Residuals) Test For a Scalar Measurement ($p = 1$)



Given the innovations signal $e(n)$

We define the scalar WSSR test statistic at time index n :

$$\gamma(n) = \sum_{j=n-W+1}^n \frac{e^2(j)}{V(j)}, \quad \text{for } n \geq W$$

Note: We estimate WSSR over a finite sliding window of length W samples.

Where:

$$V(n) = \frac{1}{W} \sum_{j=n-W+1}^n [e^2(j) - \bar{e}(j)]^2, \quad \text{for } n \geq W$$

Sample variance
over the sliding window

$$\bar{e}(n) = \frac{1}{W} \sum_{j=n-W+1}^n e(j), \quad \text{for } n \geq W$$

Sample mean
over the sliding window

Define the WSSR Hypothesis Test



By defining a threshold (later), the WSSR test becomes:

$$\text{If } \gamma(n) \begin{matrix} \geq H_1 \\ < H_0 \end{matrix} \tau, \quad (\tau = \text{Decision Threshold})$$

Read this as follows:

If $\gamma(n) \geq \tau$, then H_1 is true

If $\gamma(n) < \tau$, then H_0 is true

WSSR Test

For a scalar measurement ($p = 1$) (Continued)



For the null hypothesis H_0 , the WSSR is chi square distributed:

$$\gamma(n) \sim \chi^2(W)$$

However, for $W > 30$, the WSSR is approximately normally distributed:

$$\gamma(n) \sim N(W, 2W)$$

At the significance level α , the probability of rejecting the null Hypothesis (detecting a jump) is:

$$P\left(\left|\frac{\gamma(n) - W}{\sqrt{2W}}\right| > \left|\frac{\tau - W}{\sqrt{2W}}\right|\right) = \alpha$$

WSSR Hypothesis Test (Continued)



At the significance level α , we can create a confidence interval test:

$$\text{For } H_0: \quad P[\gamma(n) < \tau] = 1 - \alpha = .95$$

$$\text{For } H_1: \quad P[\gamma(n) \geq \tau] = \alpha = .05$$

For a significance level $\alpha = .05$, the threshold is:

$$\tau = W + 1.96\sqrt{2W}$$